

AI4EIC

Computing Frontiers Summary

Gabe Perdue and Cris Fanelli

September 10, 2021

Live notes:

<https://docs.google.com/document/d/18-IHRK43APHDOUhpXW74VRvu4tD4zBhqyqQ7VmT13z0/edit?usp=sharing>

FRIDAY, 10 SEPTEMBER

10:00 → 13:00

Day 4 morning

Conveners: Gabriel Perdue (Fermilab) , Olivier Pfister (U. Va.) , Wouter Deconinck

10:00

Computing Frontiers: Introduction

Speakers: Gabriel Perdue (Fermilab) , Olivier Pfister (U. Va.) , Wouter Deconinck (UManitoba)

5m

10:05

Quantum Simulations

Speaker: Henry Lamm (FNAL)

Lamm_Presentatio...

22m

10:27

AI/ML for complex systems

Speaker: Malachi Schram

SchramAIComplex...

22m

10:49

AI on a chip

Speaker: Farah Fahim (FNAL)

AI_on_chip_EIC.pdf

22m

11:11

Heterogeneous computing

Speaker: Travis Humble (ORNL)

Computational Fro...

22m

11:45

break

15m

12:00

Computing Frontiers - Panel Discussion

Panel discussion

Speakers: Farah Fahim (FNAL) , Henry Lamm (FNAL) , Malachi Schram, Travis Humble (ORNL)

45m

45m

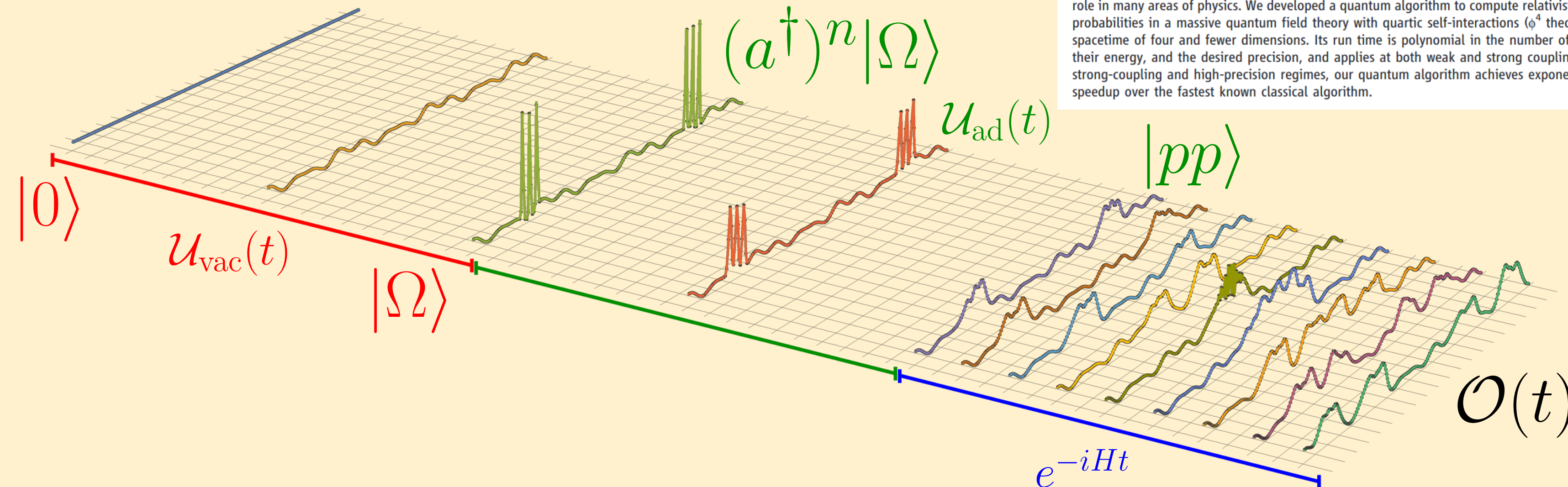
12:45

Discussion

15m

The following slides have been “borrowed” from the fantastic presentations! Thanks so much to Hank, Malachi, Farah, and Travis.

What might a galactic algorithm look like?



Quantum Algorithms for Quantum Field Theories

Stephen P. Jordan,^{1*} Keith S. M. Lee,² John Preskill³

Quantum field theory reconciles quantum mechanics and special relativity, and plays a central role in many areas of physics. We developed a quantum algorithm to compute relativistic scattering probabilities in a massive quantum field theory with quartic self-interactions (ϕ^4 theory) in spacetime of four and fewer dimensions. Its run time is polynomial in the number of particles, their energy, and the desired precision, and applies at both weak and strong coupling. In the strong-coupling and high-precision regimes, our quantum algorithm achieves exponential speedup over the fastest known classical algorithm.

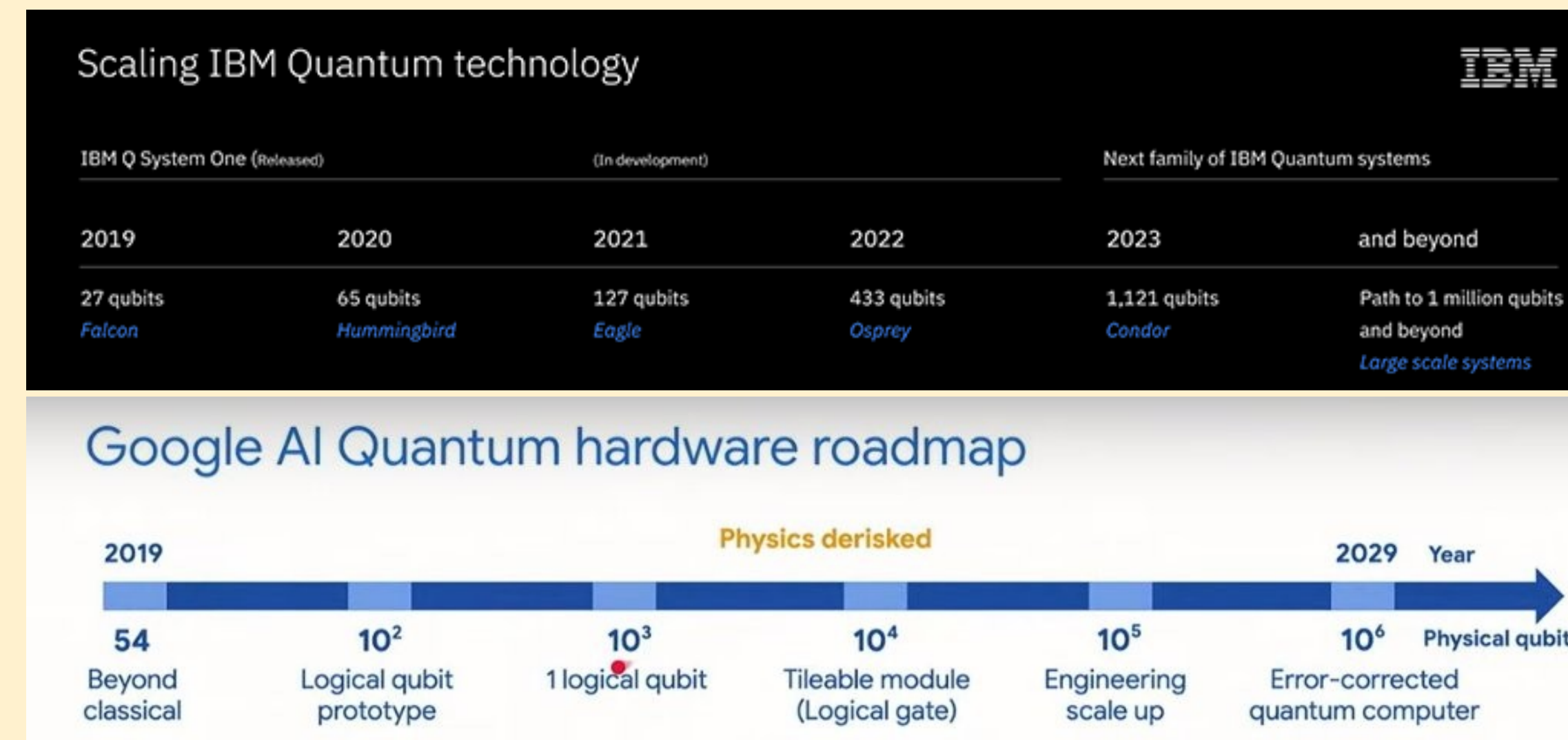
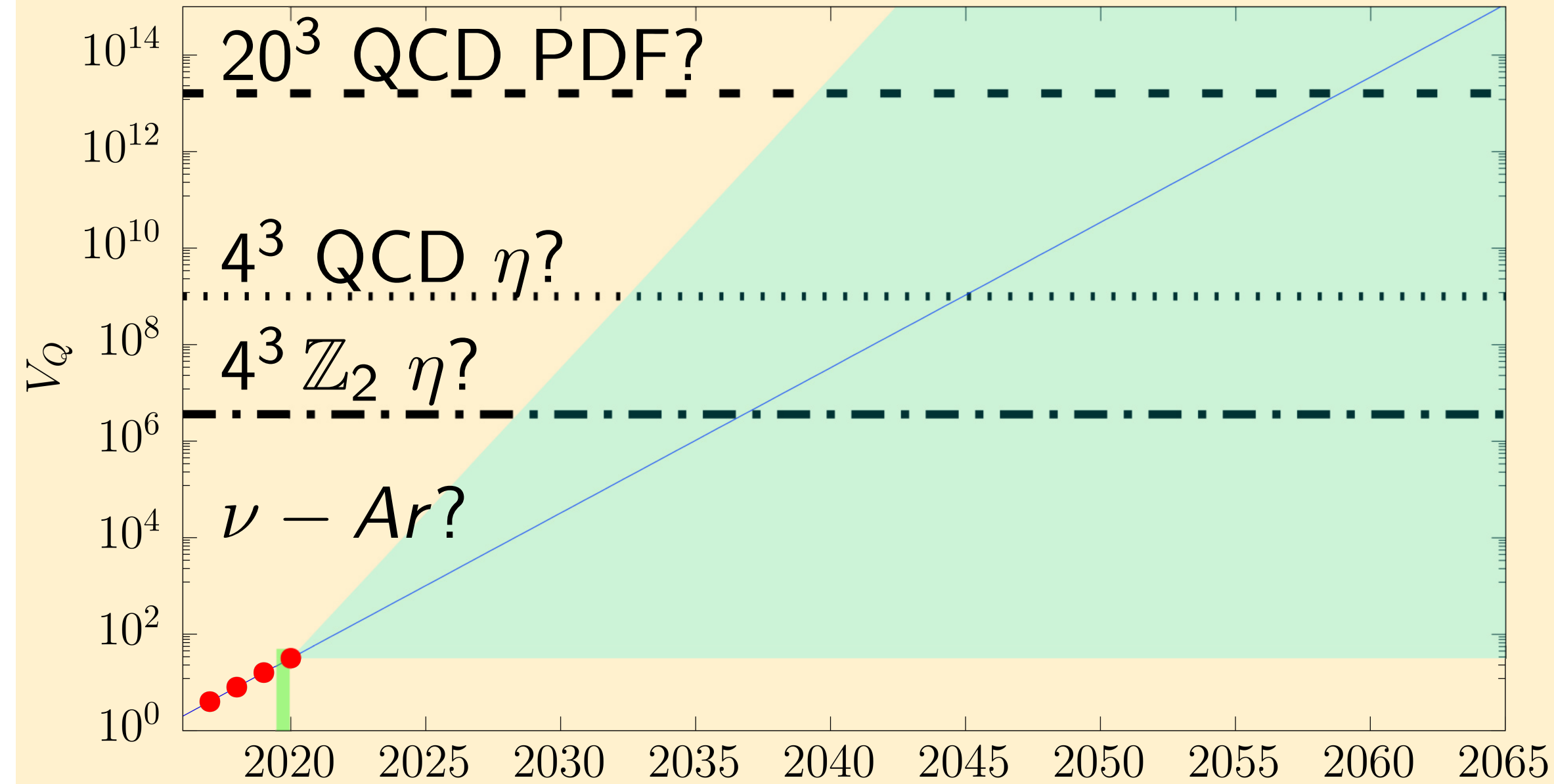
Vacuum Prep + **Adiabatic evolution** + **Trotterization** + Measurements^[2]

Example: $|\langle p\bar{p}|U(t)|\pi\pi\pi\pi\rangle|^2$ needs $\mathcal{O}(10^7)$ **logical qubits**

$$\approx \left(\frac{3 \text{ fm}}{0.05 \text{ fm}}\right)^3 \times (3 \text{ links} \times 11 \text{ qubits} + 3 \text{ colors} \times 2 \text{ flavors} \times 2 \text{ spins} \times 1 \text{ qubit})$$

[2] Jordan, S. P., K. S. M. Lee, and J. Preskill. In: *Science* 336 (2012). arXiv: 1111.3633 [quant-ph].

Where could we be when EIC runs?



It's time to go

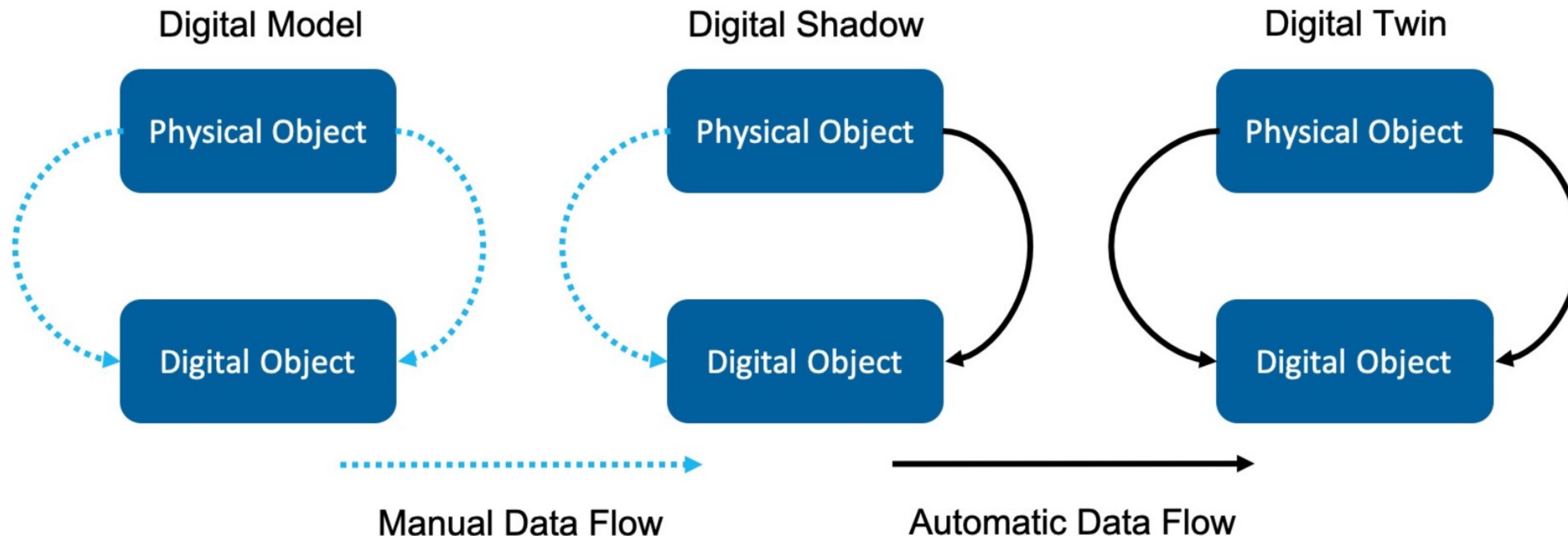
So many things to do!...and lots can be done before the machine exists

- Digitizing SU(3)
 - **Spectroscopy** for approximations
 - Explicit **circuits**
- Reducing the errors
 - e.g. Finite volume, finite a , a_t , decimation errors, fidelity to obtain **realistic** resource estimates
- Algorithms for **state prep, smearing**
- Investigate desirable properties
 - **PDF?, Viscosity?**
- **Actual** simulations of toy models
 - \mathbb{Z}_2 & \mathbb{D}_4

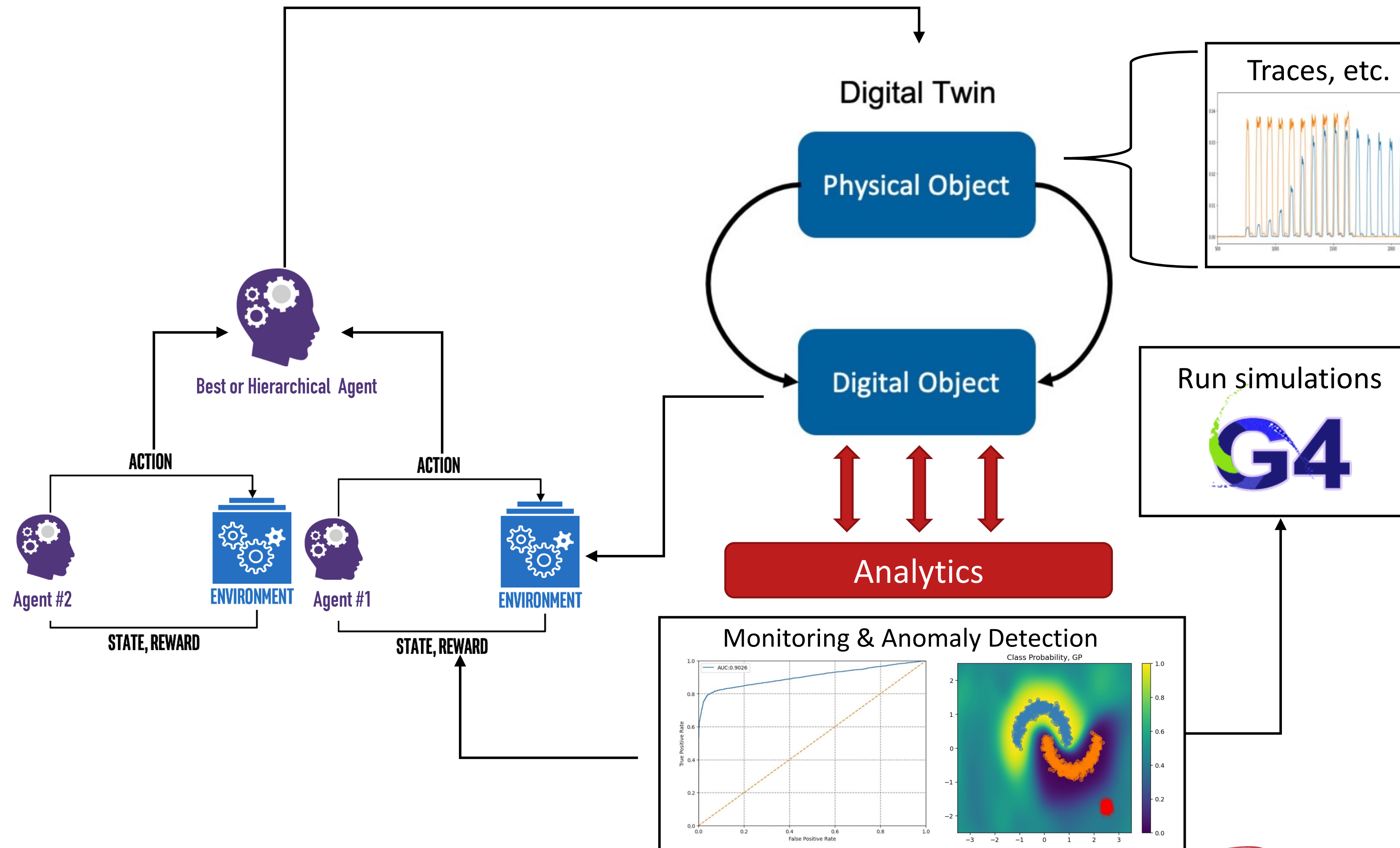


Digital twin definitions

- **Digital Model:** a digital version of a pre-existing or planned physical object
- **Digital Shadow:** digital representation of a physical object with a one-way data flow from the physical to digital object
- **Digital Twin:** data flows between a physical object and a digital object are fully integrated and bilateral

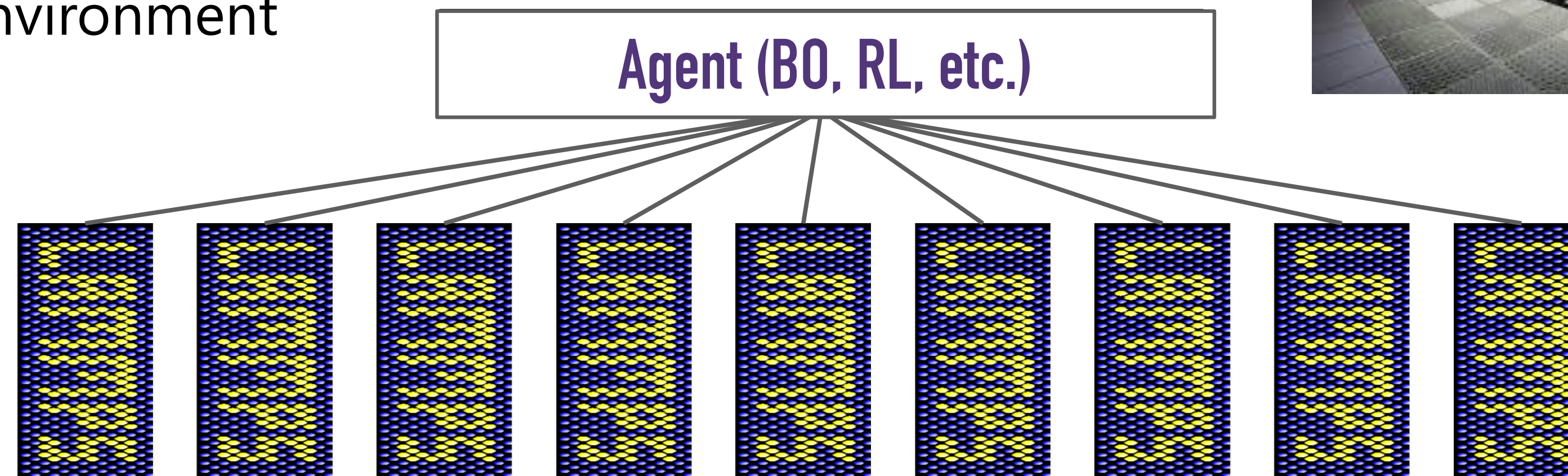
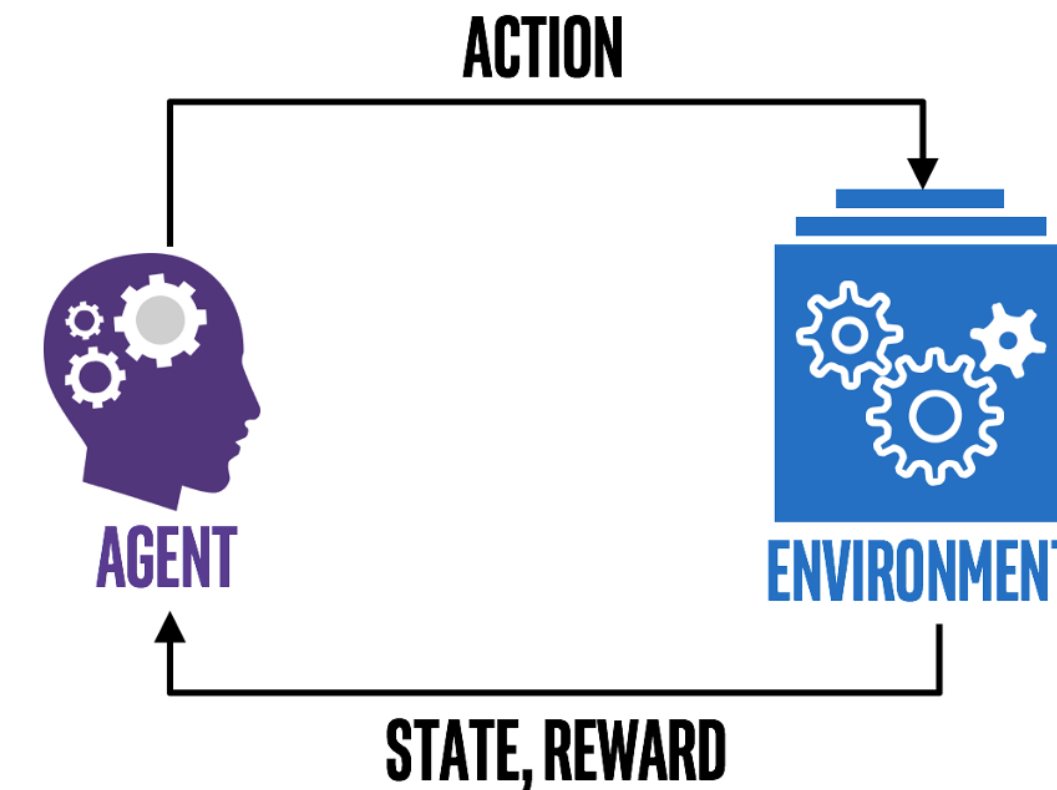


Extending digital twin and analytics for control workflow

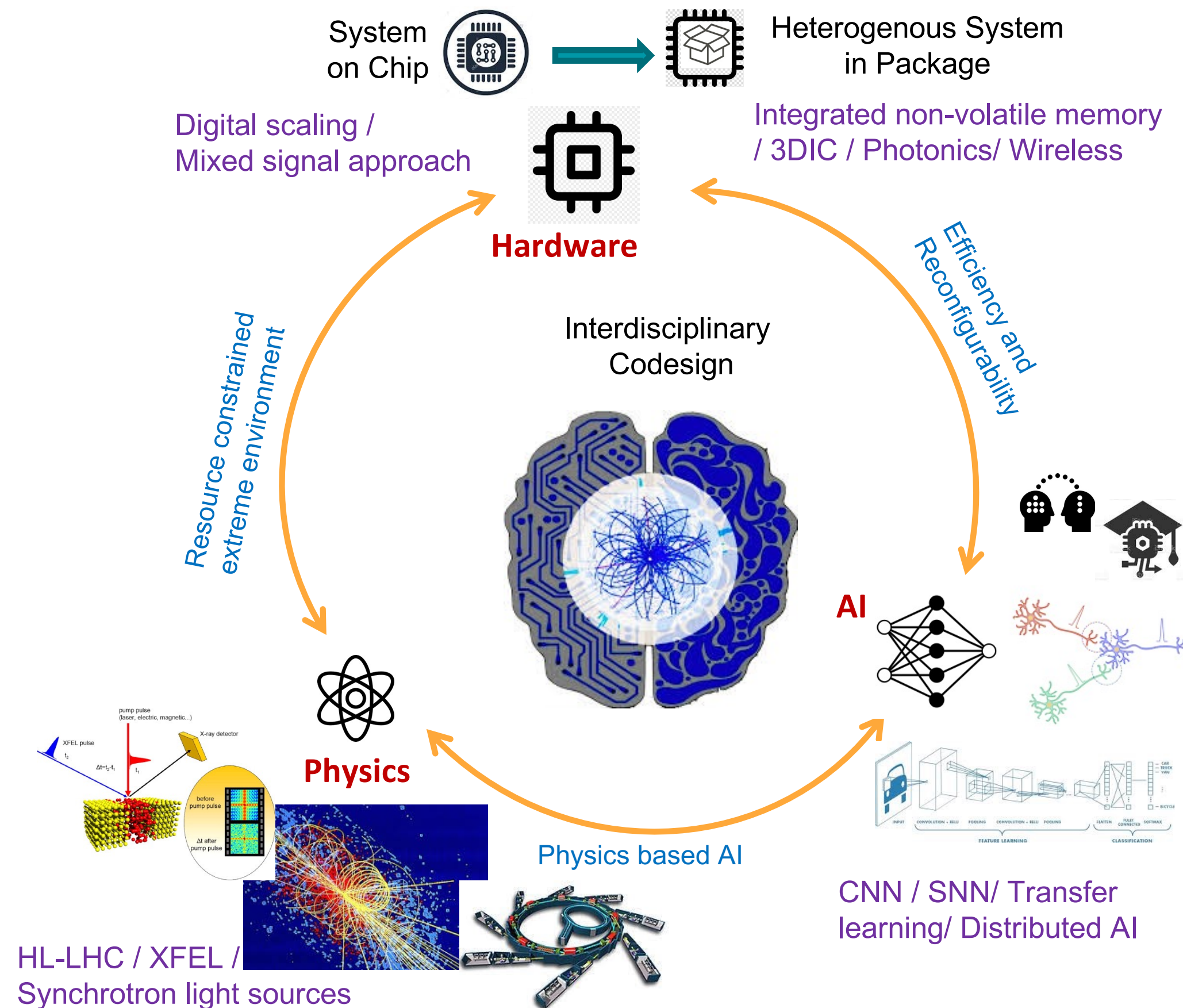


Scaling workflow on HPC system

- Digital Object is interfaced with industry standard OpenAI gym environment
- To accelerate the data generation we developed a MPI-based framework
- We created an agent that maps the action-reward for all simulations
- A production job split the MPI communications between the agent and each Digital Object environment

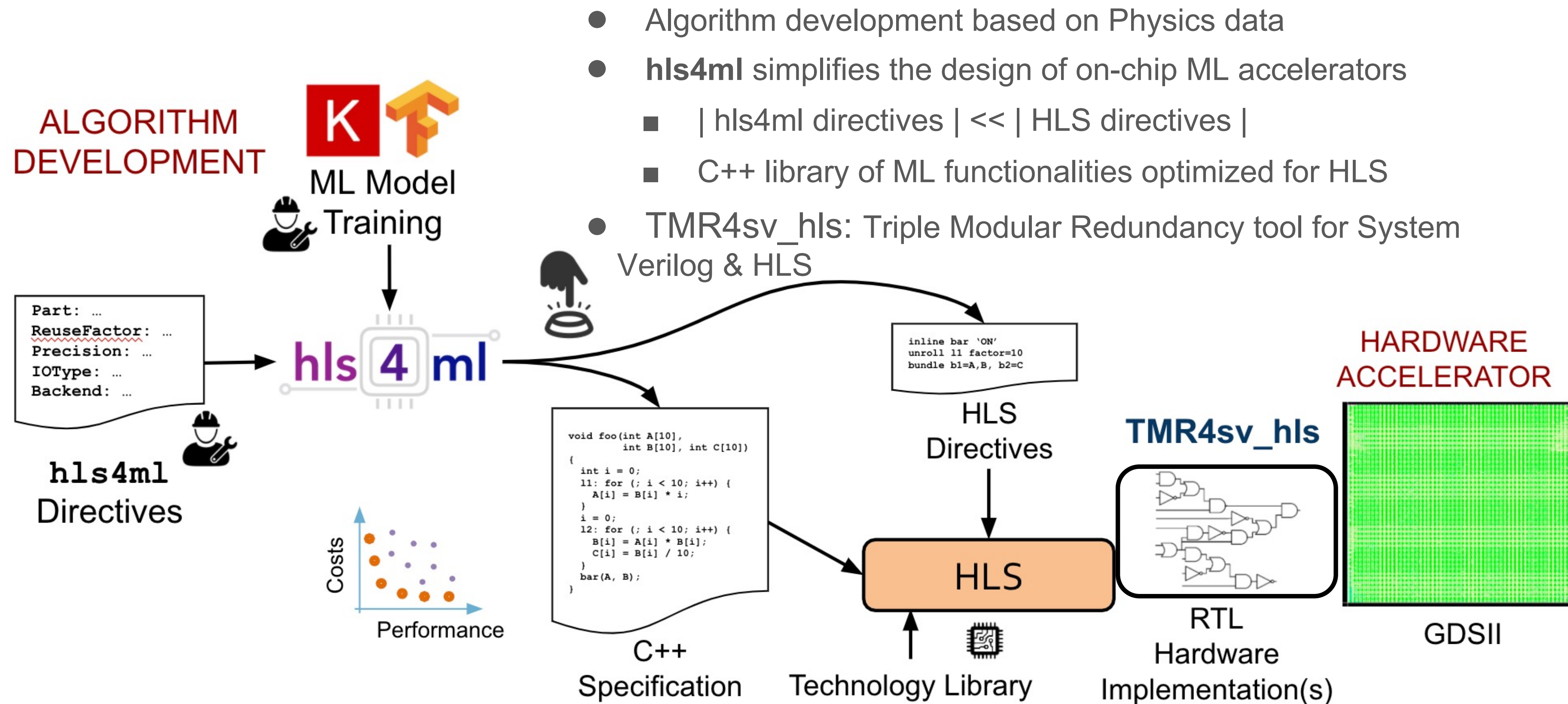


On-detector intelligence using on-chip Machine Learning

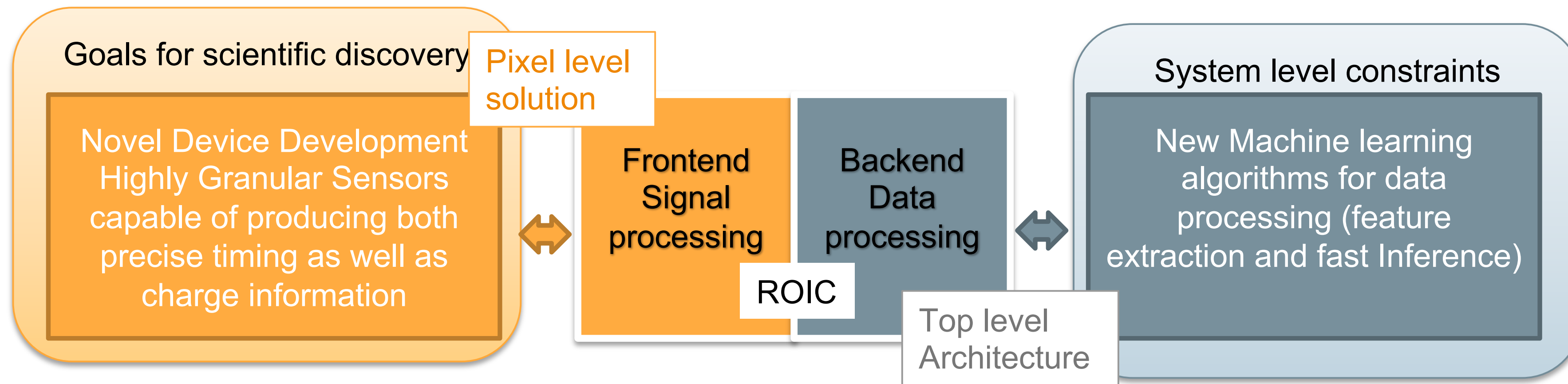


- Resource constrained environment
 - High Radiation
 - Limited Power/Material budget
 - Where should this intelligence be added
- Efficiency and reconfigurability
 - Ultra-low energy per inference at extremely high rates (10's ns)
 - Reprogram both network and parameters
 - On-chip learning / inference
- Physics based Algorithms
 - Independent events
 - Depth vs. classification

Physics Driven Hardware Co-design



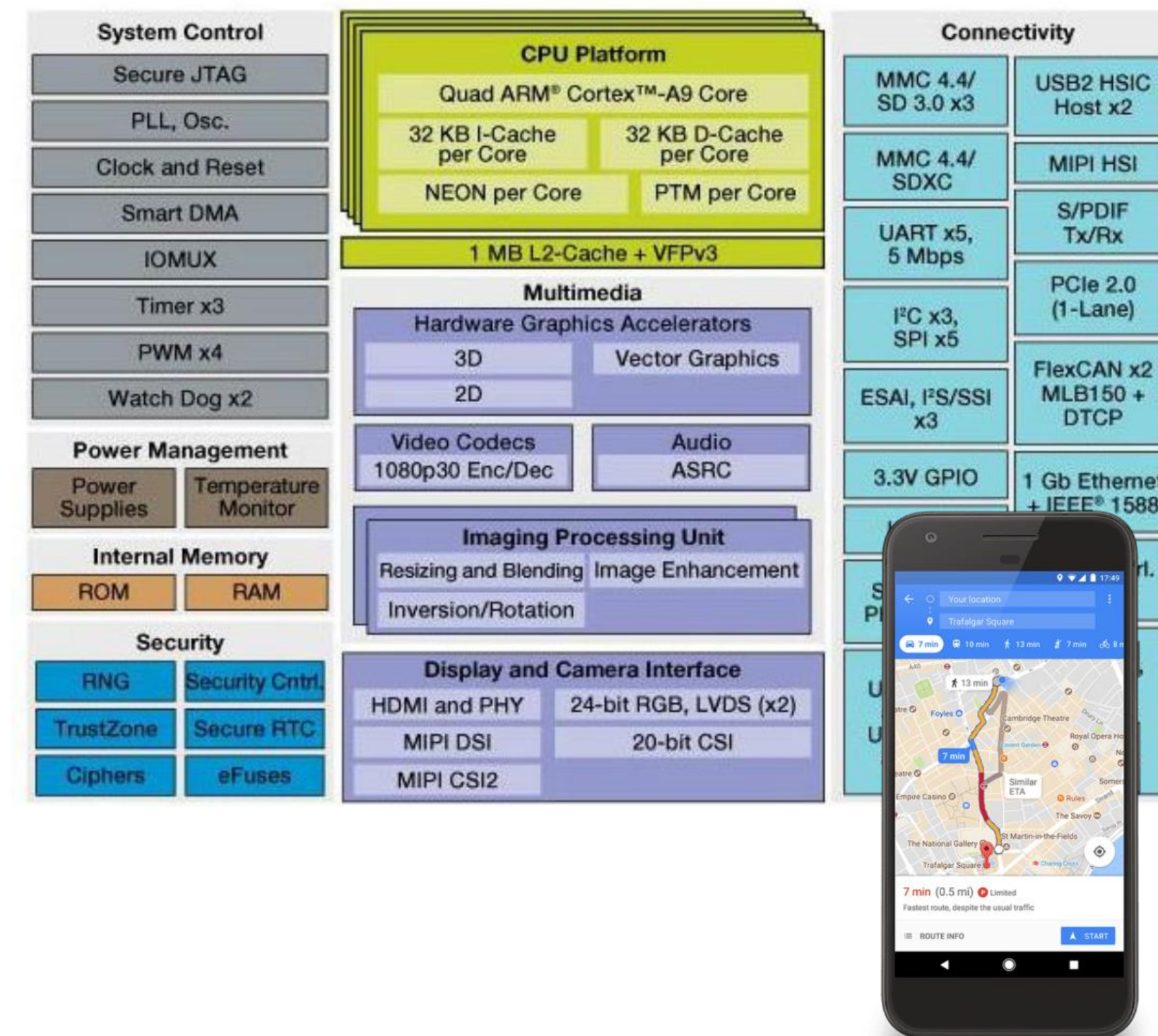
Co-design with algorithm



- Convert raw data to physics information
- Reconfigurable pixel clusters for classification dependent on detector geometries
- Create hierarchical network and enable parallel computation.

Heterogenous Computing

- The only tool left to a computer architect for extracting continued performance improvements is to use transistors more efficiently by specializing the architecture to the target scientific problem
- Computer vendors are pursuing systems built from combinations of different types of processors to improve capabilities, boost performance, and meet energy efficiency goals.



Productive Computational Science in the Era of Extreme Heterogeneity <https://doi.org/10.2172/1473756>

Five Priority Research Directions

1. Maintaining and Improving Programmer Productivity

- *Flexible, expressive, programming models and languages*
- *Intelligent, domain-aware compilers and software development tools*
- *Composition of disparate software component content*

2. Managing System Resources Intelligently

- *Automated methods using introspection and machine learning*
- *Optimize for performance, energy efficiency, and availability*

3. Modeling and Predicting Performance

- *Evaluate impact of potential system designs and application mappings*
- *Model-automated optimization of applications*

4. Enabling Reproducible Science Despite Diverse Processors and Non-Determinism

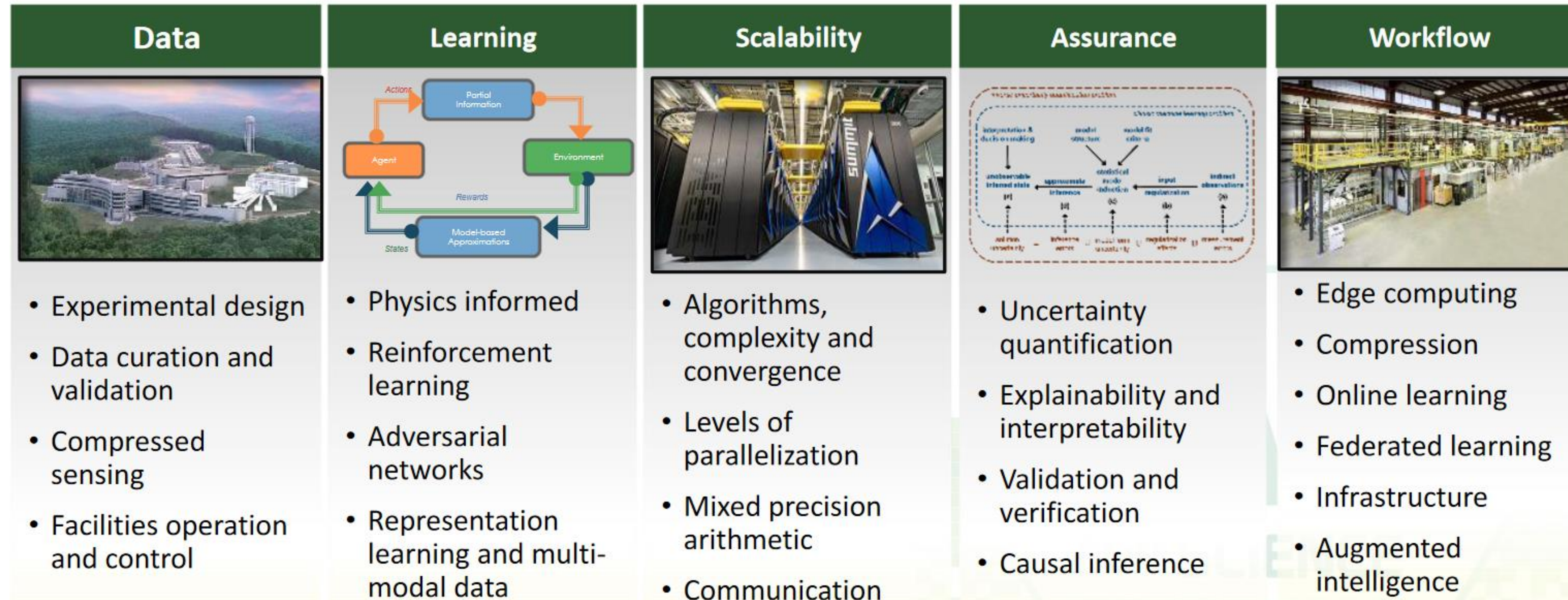
- *Methods for validation on non-deterministic architectures*
- *Detection and mitigation of pervasive faults and errors*

5. Facilitating Data Management, Analytics, and Workflows

- *Mapping a science workflow to heterogeneous hardware and software services*
- *Adapting workflows and services through machine learning approaches*

Productive Computational Science in the Era of Extreme Heterogeneity <https://doi.org/10.2172/1473756>

Research Challenges in Heterogenous AI



Panel discussion

Wide-ranging conversation that touched on:

- * effective models for maintaining partnerships with industry 10 years from now,
- * upcoming changes to the “boundaries” between different steps of the computing workflow - tasks that live in one bucket now will likely move to another!,
- * the ways quantum simulation may guide data analysis,
- * how we will make advanced platforms available,
- * what will be important 10 years from now but isn’t part of the discussion today,
- * and even what the water supply needs to look like when locating your datacenters!

See the live doc for details!